#### CNN reconstruction of air shower events with TA FD

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#### Motivation

- Currently, FD is analyzed by the IMC method.
- IMC method is computationally time-consuming due to many of MC simulation.
- In the existing reconstruction, only one-dimensional parameters (such as  $X_{\text{max}}$ ) can be obtained



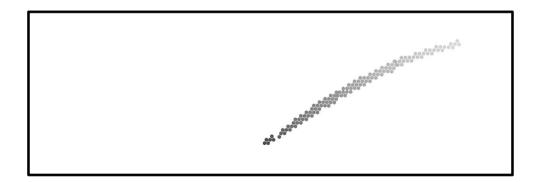
Development of reconstruction method by deep learning using FD event display as training data

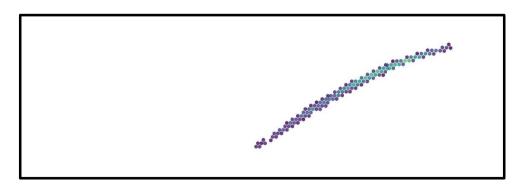
### Research method

- 1. Air shower simulation to create pseudo data in the format as the actual data
- 2. Create event display from the created pseudo data
- 3. Create datasets for training from event display
- 4. Use the dataset to train the model to learn the features
- 5. Estimate air shower parameters from learned features L. Zenith, Azimuth, CoreX, CoreY, Energy, Xmax
- 6. Evaluate the model by comparing the estimated values with the true

### Event display

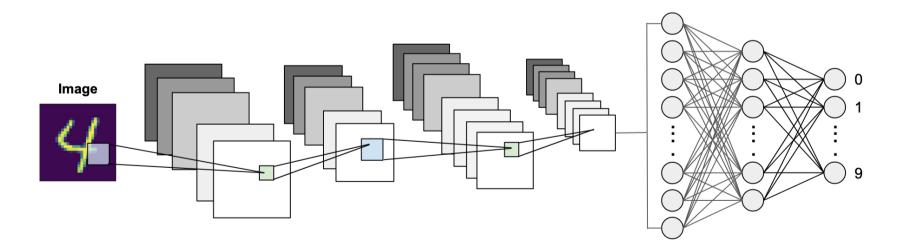
- event display images (96×32)
- The color map shows the passage of time, from blue to red
- The number of PMTs that detected signals is 15 or more, and the total number of PMTs detecting signals within each PMT camera is 128 or fewer





#### Model

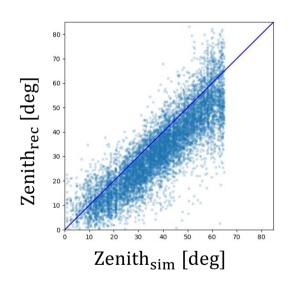
- Using Python library : TensorFlow.Keras
- Using Model: CNN( Convolution Neural Network)
  - -Models specialized for learning with images
  - -Use convolutional and pooling layers for feature extraction

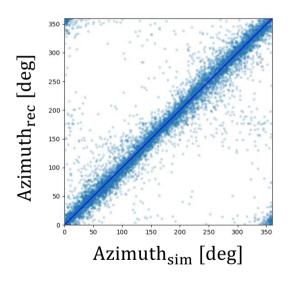


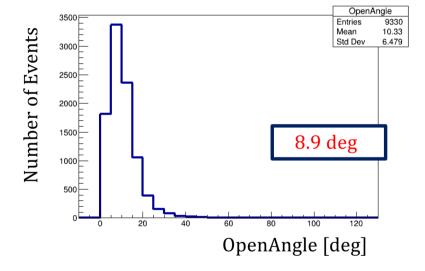
#### **Previous Status**

- geometry estimation results are as accurate as conventional reconstruction methods.
- No energy & Xmax estimation is possible.
- To add energy & Xmax, I implemented a multi-input model to increase the number of features.
- Analysis with 6 parameters allowed learning of the core position, but failed to learn Zenith, Azimuth, energy, and Xmax.

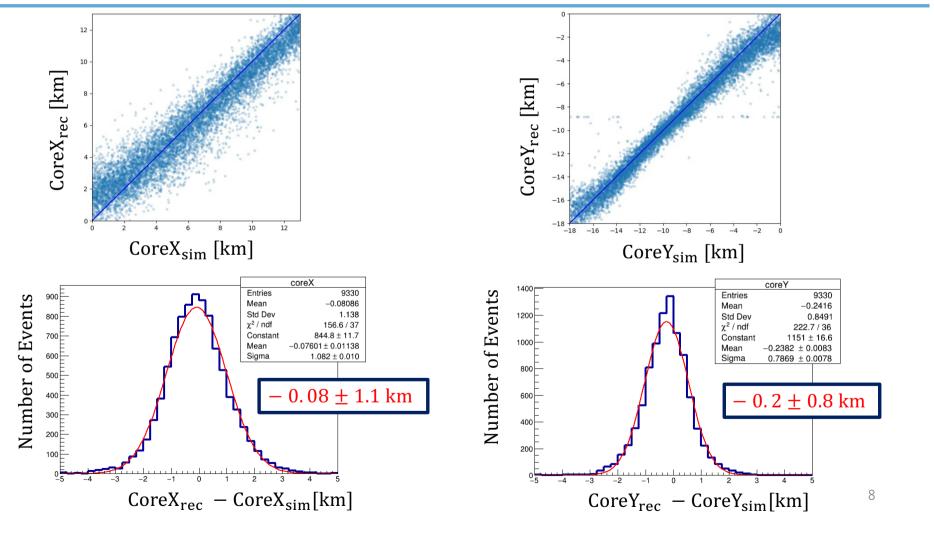
## Analysis Results (Zenith, Azimuth)



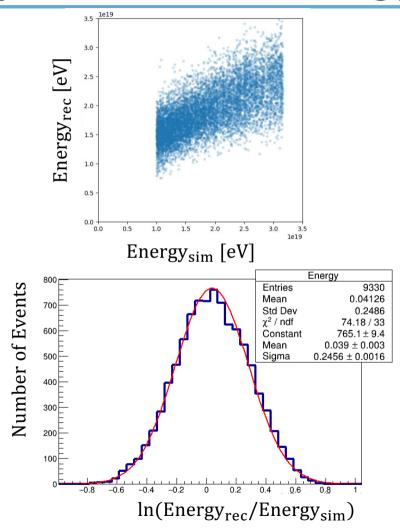


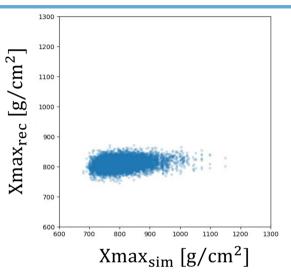


# Analysis Results (Core)



# Analysis Results (Energy, Xmax)





## Summary & Next to do

# Summary

- Debugging work was performed.
- The system has returned to its expected behavior, enabling parameter analysis.

### Next to do

- Improve the model so that Xmax can learn.
- Prepare training data using CONEX and create a training model.

Back up

## **Simulation Conditions**

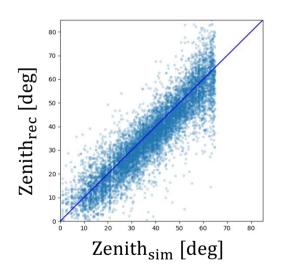
Training dataset	
19.0 ~ 19.1	10000 event
19.1 ~ 19.2	10000 event
19.2 ~ 19.3	10000 event
19.3 ~ 19.4	10000 event
19.4 ~ 19.5	10000 event
Event display create before	50000 event
create after	46762 event
Test dataset	
Test dataset $19.0 \sim 19.1$	2000 event
	2000 event 2000 event
19.0 ~ 19.1	
19.0 ~ 19.1 19.1 ~ 19.2	2000 event
$19.0 \sim 19.1$ $19.1 \sim 19.2$ $19.2 \sim 19.3$	2000 event 2000 event
$19.0 \sim 19.1$ $19.1 \sim 19.2$ $19.2 \sim 19.3$ $19.3 \sim 19.4$	2000 event 2000 event 2000 event

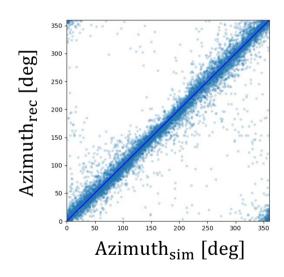
9330 event

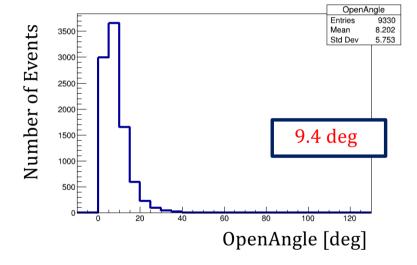
create after

particle type	Proton
FD Station	BR
Energy (log(E/eV))	19.0 ~ 19.5
Zenith	0 ~ 65
Azimuth	0 ~ 360
Core X (km)	0 ~ 13
Core Y (km)	<b>−</b> 18 ~ 0
coordinate origin	CLF
Missing energy	CORSIKA2009
Hadron interaction model	QGSJET-II-04

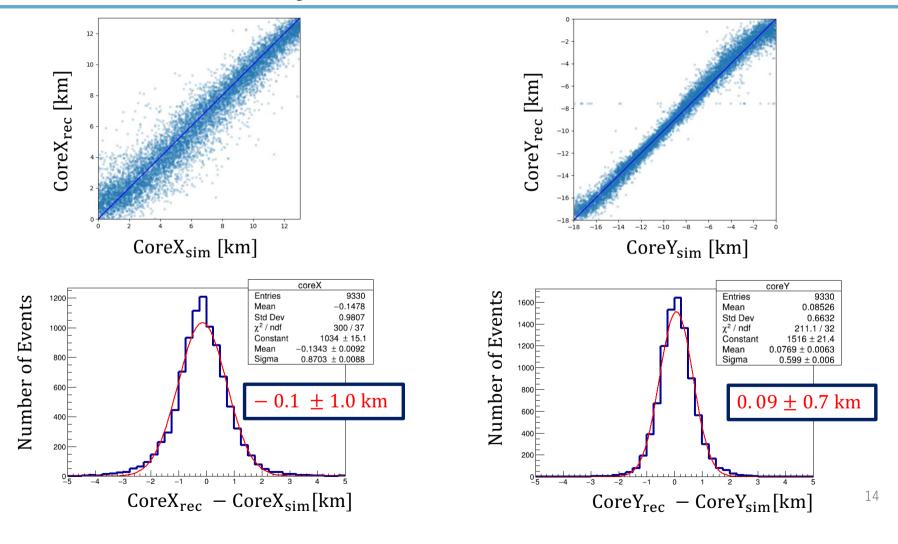
### 4Para Model Analysis Results (Zenith, Azimuth)







## 4Para Model Analysis Results (Core)



### 6Parameter Model

